

Homework 2

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1

we know that

$$x_1 = 2$$

$$x_2 = 3$$

$$\frac{\partial L}{\partial o} = 5$$

$$o = x_1 * x_2$$

To find $\frac{\partial L}{\partial x_1}$ and $\frac{\partial L}{\partial x_2}$ we use the Chain rule which gives us $\frac{\partial L}{\partial o} \frac{\partial o}{\partial x_1}$ and $\frac{\partial L}{\partial o} \frac{\partial o}{\partial x_2}$ respectively. It can be derived that $\frac{\partial o}{\partial x_1}$ and $\frac{\partial o}{\partial x_2}$ are x_2 and x_1 respectively

Therefore we can solve for both

$$\begin{aligned}\frac{\partial L}{\partial x_1} &= \frac{\partial L}{\partial o} \frac{\partial o}{\partial x_1} \\ &= 5 * x_2 \\ &= 5 * 3\end{aligned}$$

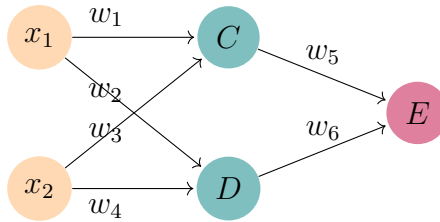
$$\frac{\partial L}{\partial x_1} = 15$$

$$\begin{aligned}\frac{\partial L}{\partial x_2} &= \frac{\partial L}{\partial o} \frac{\partial o}{\partial x_2} \\ &= 5 * x_1 \\ &= 5 * 2\end{aligned}$$

$$\frac{\partial L}{\partial x_2} = 10$$

2

The Neural Network is as follows.



We know that $w_1 = 0.1$, $w_2 = 0.5$, $w_3 = 0.4$, $w_4 = 0.3$, $w_5 = 0.2$, $w_6 = 0.6$. The Hidden Layer and Output Layer, $y_h()$ and $y_o()$ respectively, both have the activation function of $y_n(z) = \frac{1}{1+e^{-z}}$. The Loss function is $L = \frac{1}{2}(y - \hat{y})^2$, where y is the expected value and \hat{y} is the actual value

Our starting point is $\left(\begin{bmatrix} 0.82 \\ 0.23 \end{bmatrix} \quad 0 \right)$. The Weights for each nodes are as follows

$$\vec{w}_C = \begin{bmatrix} w_1 \\ w_3 \end{bmatrix}$$

$$\vec{w}_D = \begin{bmatrix} w_2 \\ w_4 \end{bmatrix}$$

$$\vec{w}_E = \begin{bmatrix} w_5 \\ w_6 \end{bmatrix}$$

The Weights for the Hidden and Output layer as as follow

$$\begin{aligned} W_h &= \begin{bmatrix} w_C & w_D \\ w_1 & w_2 \\ w_3 & w_4 \end{bmatrix} & W_o &= \begin{bmatrix} w_E \\ w_5 \\ w_6 \end{bmatrix} \end{aligned}$$

The z_C , z_D and z_E can be written as followed

$$z_C = w_1 * x_1 + w_3 * x_2 \quad z_D = w_2 * x_1 + w_4 * x_2 \quad z_E = w_5 * y_C + w_6 * y_D$$

For solving forward propogation we do the following steps

$$W_h^T * \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} z_C \\ z_D \end{bmatrix} \quad (1)$$

$$\begin{bmatrix} y_n(z_C) \\ y_n(z_D) \end{bmatrix} = \begin{bmatrix} y_C \\ y_D \end{bmatrix} \quad (2)$$

$$W_o^T * \begin{bmatrix} y_C \\ y_D \end{bmatrix} = z_E \quad (3)$$

$$y_E = y_n(z_E) \quad (4)$$

Plugging in the values we get this

$$\begin{aligned} \begin{bmatrix} 0.1 & 0.5 \\ 0.4 & 0.3 \end{bmatrix}^T * \begin{bmatrix} 0.82 \\ 0.23 \end{bmatrix} &= \begin{bmatrix} 0.174 \\ 0.479 \end{bmatrix} \\ \begin{bmatrix} y_h(0.174) \\ y_h(0.479) \end{bmatrix} &= \begin{bmatrix} 0.5433906 \\ 0.6175177 \end{bmatrix} \\ \begin{bmatrix} 0.2 \\ 0.6 \end{bmatrix}^T * \begin{bmatrix} 0.5433906 \\ 0.6175177 \end{bmatrix} &= 0.47918874 \\ y_o(0.47918874) &= 0.617556289 \end{aligned}$$

The ending value, y_E , is 0.617556289 and the other values are as follows

$$\begin{aligned} y_C &= 0.5433906 & y_D &= 0.6175177 \\ z_C &= 0.174 & z_D &= 0.479 & z_E &= 0.471988174 \end{aligned}$$

For Back propagation, the outer layer weights will derived as

$$\begin{aligned} \frac{\partial L}{\partial w_5} &= \frac{\partial L}{\partial y_e} * \frac{\partial y_e}{\partial z_e} * \frac{\partial z_e}{\partial w_5} & \frac{\partial L}{\partial w_6} &= \frac{\partial L}{\partial y_e} * \frac{\partial y_e}{\partial z_e} * \frac{\partial z_e}{\partial w_6} & \delta_o &= \frac{\partial L}{\partial y_e} * \frac{\partial y_e}{\partial z_e} \\ \frac{\partial L}{\partial w_5} &= \delta_o * \frac{\partial z_e}{\partial w_5} & \frac{\partial L}{\partial w_6} &= \delta_o * \frac{\partial z_e}{\partial w_6} \end{aligned}$$

For the weights that connect to Node C, we find the rate of change as

$$\begin{aligned} \frac{\partial L}{\partial w_1} &= \frac{\partial L}{\partial y_c} * \frac{\partial y_c}{\partial z_c} * \frac{\partial z_c}{\partial w_1} & \frac{\partial L}{\partial w_3} &= \frac{\partial L}{\partial y_c} * \frac{\partial y_c}{\partial z_c} * \frac{\partial z_c}{\partial w_3} & \frac{\partial L}{\partial y_c} &= \frac{\partial L}{\partial y_e} * \frac{\partial y_e}{\partial z_e} * \frac{\partial z_e}{\partial y_c} \\ & & & & &= \delta_o * \frac{\partial z_e}{\partial y_c} \end{aligned}$$

For the weights that connect to Node D, we find the rate of change as

$$\begin{aligned}\frac{\partial L}{\partial w_2} &= \frac{\partial L}{\partial y_d} * \frac{\partial y_d}{\partial z_d} * \frac{\partial z_d}{\partial w_2} & \frac{\partial L}{\partial w_4} &= \frac{\partial L}{\partial y_d} * \frac{\partial y_d}{\partial z_d} * \frac{\partial z_d}{\partial w_4} & \frac{\partial L}{\partial y_d} &= \frac{\partial L}{\partial y_e} * \frac{\partial y_e}{\partial z_e} * \frac{\partial z_e}{\partial y_d} \\ & & & & &= \delta_o * \frac{\partial z_e}{\partial y_d}\end{aligned}$$

Derivatives for L, y are as follow

$$\frac{\partial L}{\partial y_e} = -(y - y_e) \qquad \frac{\partial y_n}{\partial z_n} = y_n(z_n) * (1 - y_n(z_n))$$

Now lets solve for each weights

Weights w_5 and w_6

$$\begin{aligned}\delta_o &= \frac{\partial L}{\partial y_e} * \frac{\partial y_e}{\partial z_e} \\ &= -(0 - y_E) * (y_o(z_E) * (1 - y_o(z_E))) \\ &= 0.09032574 \\ \frac{\partial L}{\partial w_5} &= \delta_o * \frac{\partial z_e}{\partial w_5} & \frac{\partial L}{\partial w_6} &= \delta_o * \frac{\partial z_e}{\partial w_6} \\ &= \delta_o * y_C & &= \delta_o * y_D \\ &= 0.09032574 * 0.5433906 & &= 0.09032574 * 0.6175177 \\ &= 0.049082158 & &= 0.055777743\end{aligned}$$

Weights w_1 and w_3

$$\begin{aligned}\frac{\partial L}{\partial y_C} &= \delta_o * \frac{\partial z_e}{\partial y_C} \\ &= \delta_o * w_5 \\ &= 0.09032574 * 0.2 \\ &= 0.018065148 \\ \frac{\partial L}{\partial w_1} &= \frac{\partial L}{\partial y_C} * \frac{\partial y_C}{\partial z_C} * \frac{\partial z_C}{\partial w_1} & \frac{\partial L}{\partial w_3} &= \frac{\partial L}{\partial y_C} * \frac{\partial y_C}{\partial z_C} * \frac{\partial z_C}{\partial w_3} \\ &= 0.018065148 * (y_h(z_C) * (1 - y_h(z_C))) * x_1 & &= 0.018065148 * (y_h(z_C) * (1 - y_h(z_C))) * x_2 \\ &= 0.018065148 * 0.248117255 * 0.82 & &= 0.018065148 * 0.248117255 * 0.23 \\ &= 0.00367546544 & &= 0.00103092323\end{aligned}$$

Weights w_2 and w_4

$$\frac{\partial L}{\partial y_D} = \delta_o * \frac{\partial z_e}{\partial y_D}$$

$$= \delta_o * w_6$$

$$= 0.09032574 * 0.6$$

$$= 0.05419544$$

$$\frac{\partial L}{\partial w_2} = \frac{\partial L}{\partial y_D} * \frac{\partial y_D}{\partial z_D} * \frac{\partial z_D}{\partial w_2}$$

$$= 0.05419544 * (y_h(z_D) * (1 - y_h(z_D))) * x_1$$

$$= 0.05419544 * 0.23618959 * 0.82$$

$$= 0.01049632697$$

$$\frac{\partial L}{\partial w_4} = \frac{\partial L}{\partial y_D} * \frac{\partial y_D}{\partial z_D} * \frac{\partial z_D}{\partial w_4}$$

$$= 0.05419544 * (y_h(z_D) * (1 - y_h(z_D))) * x_2$$

$$= 0.05419544 * 0.23618959 * 0.23$$

$$= 0.00294409171$$

The Final $\frac{\partial L}{\partial \vec{w}}$ is as follows

$$\frac{\partial L}{\partial \vec{w}} = \begin{bmatrix} 0.00367546544 \\ 0.01049632697 \\ 0.00103092323 \\ 0.00294409171 \\ 0.049082158 \\ 0.055777743 \end{bmatrix}$$

The learning rate ϵ is 0.7 so the new weights are as followed

$$\begin{aligned} \vec{w}' &= \vec{w} - \epsilon * \frac{\partial L}{\partial \vec{w}} \\ &= \begin{bmatrix} 0.1 \\ 0.5 \\ 0.4 \\ 0.3 \\ 0.2 \\ 0.6 \end{bmatrix} - 0.7 * \begin{bmatrix} 0.00367546544 \\ 0.01049632697 \\ 0.00103092323 \\ 0.00294409171 \\ 0.049082158 \\ 0.055777743 \end{bmatrix} \\ &= \begin{bmatrix} 0.09742717419 \\ 0.49265257112 \\ 0.39927835373 \\ 0.2979391358 \\ 0.1656424894 \\ 0.560955799 \end{bmatrix} \end{aligned}$$